

## Structured Singular Value $\mu$

We know that it is generally harder to destabilize the MD loop with structured perturbation than unstructured perturbation

Definition: The structured singular value  $\mu(M)$  is defined as:

$$\frac{1}{\mu(M)} = \min_{\Delta} \{ \sigma_{\max}(\Delta) \mid \det(I - M\Delta) = 0 \}, \text{ structured } \Delta$$

Generally speaking,  $M$  is a complex valued matrix

Thus, the smaller  $\mu(M)$  is, the better for robustness. Let us consider two extreme cases.

### Least Restrictive Structure

If  $\Delta$  is complex and without any structure,

$$\mu(M) = \sigma_{\max}(M)$$

Proof: Perform SVD on  $M$ :  $M = U\Sigma V^H$ , define

$$\Delta = V \begin{pmatrix} \frac{1}{\sigma_{\max}(M)} & & & \\ & 0 & & \\ & & \ddots & \\ & & & 0 \end{pmatrix} U^H, \text{ thus } \sigma_{\max}(\Delta) = \frac{1}{\sigma_{\max}(M)}$$

$$M\Delta = U \begin{pmatrix} 1 & & & \\ & 0 & & \\ & & \ddots & \\ & & & 0 \end{pmatrix} U^H = U \begin{pmatrix} 1 & & & \\ & 0 & & \\ & & \ddots & \\ & & & 0 \end{pmatrix} U^{-1}, \text{ which means that } 1 \text{ is}$$

an eigenvalue of  $M\Delta$ . Therefore  $\mu(M) \geq \sigma_{\max}(M)$

To prove the opposite, notice that if 1 is an eigenvalue of  $M\Delta$ , then  $\exists x \neq 0$

$M\Delta x = x$ . But then we have:

$$\|x\| = \|M\Delta x\| \leq \sigma_{\max}(M) \sigma_{\max}(\Delta) \|x\|, \text{ or}$$

$$\sigma_{\max}(M) \cdot \sigma_{\max}(\Delta) \geq 1$$

$$\frac{1}{\sigma_{\max}(\Delta)} \leq \sigma_{\max}(M), \text{ which implies } \mu(M) \leq \sigma_{\max}(M) \quad \text{q.e.d.}$$

### Most Restrictive Case

If  $\Delta$  is a complex scalar, then  $\mu(M) = \rho(M)$

Proof: Perform eigenvalue decomposition on  $M$ :

$$M = U \Lambda U^{-1}$$

Define  $\Delta = \frac{1}{\lambda_{\max}}$ , thus  $M\Delta$  has an eigenvalue at 1. Therefore

$$\mu(M) \geq |\lambda_{\max}| = \rho(M)$$

To prove the opposite, if 1 is an eigenvalue of  $M\Delta$ , then  $\exists x \neq 0$  such that  $M\Delta x = x$ , but since  $\Delta$  is scalar, this means that  $x$  is an eigenvector of  $M$  with eigenvalue  $\frac{1}{\Delta}$ , the smallest  $\Delta$  that has this property is  $\frac{1}{\lambda_{\max}}$

Thus for any structured singular value:

$$\rho(M) \leq \mu(M) \leq \sigma_{\max}(M)$$

Example:  $M = \begin{bmatrix} 2 & 2 \\ -1 & -1 \end{bmatrix}$

With unstructured uncertainty, we can find the smallest  $\Delta$  that makes  $\det(I - M\Delta) = 0$  as  $\Delta = \begin{bmatrix} 0.2 & -0.1 \\ 0.2 & -0.1 \end{bmatrix}$  with  $\sigma_{\max}(\Delta) = \frac{1}{\sigma_{\max}(M)} = \frac{1}{3.162}$

For scalar  $\Delta$ , we can find the smallest  $\Delta$  that makes  $\det(I - M\Delta) = 0$  as  $\Delta = \frac{1}{\lambda_{\max}(M)} = 1$ , since  $I - M = \begin{bmatrix} -1 & -2 \\ 1 & 2 \end{bmatrix} \rightarrow \det = 0$

So we expect that for any other structured uncertainty

$$1 \leq \mu(M) \leq 3.162$$

Consider the diagonal uncertainty  $\Delta = \begin{bmatrix} \Delta_1 & 0 \\ 0 & \Delta_2 \end{bmatrix}$ , we can compute

that the smallest  $\Delta$  that makes  $\det(I - M\Delta) = 0$  is

$$\Delta = \begin{bmatrix} 1/3 & 0 \\ 0 & -1/3 \end{bmatrix},$$

thus  $\sigma_{\max}(\Delta) = \frac{1}{3}$ ,  $\mu(M) = 3$

In MATLAB, use the function 'mu' to approximate  $\mu(M)$

often,  $\min_D \sigma_{\max}(DMO^{-1})$  is used as upperbound for  $\mu(M)$

If there are three or fewer blocks, this upperbound is tight  
Relation to Robust Stability with structured uncertainty

Recall that RS means:

$$\forall \omega, \forall \Delta, \det(\Sigma - M\Delta) \neq 0$$

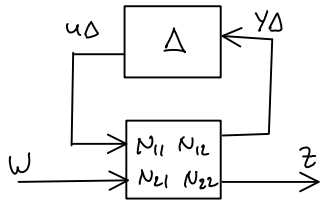
Therefore for structured uncertainty:

Thm: Assume that the plant is NS and  $\Delta$  is stable, then the  $M\Delta$  block is stable for all allowed uncertainty block  $\Delta$ ,  $\sigma_{\max}(\Delta) \leq 1$  iff  $\exists \omega, \mu(M(j\omega)) < 1$

In MATLAB, use 'ssv' to produce structured singular value plot.

### Robust Performance

We can use  $\mu$  to assess the robustness property of the performance



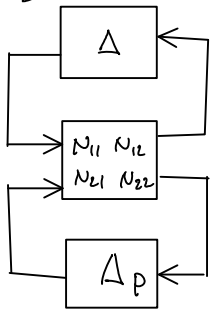
Recall that RP means:

$$NS \text{ and } \|F\|_{\infty} < 1 \text{ for all } \|\Delta\|_{\infty} \leq 1$$

where  $F$  is the TF from  $w$  to  $z$

$$F = F_u(N, \Delta)$$

Theorem: RP if and only if

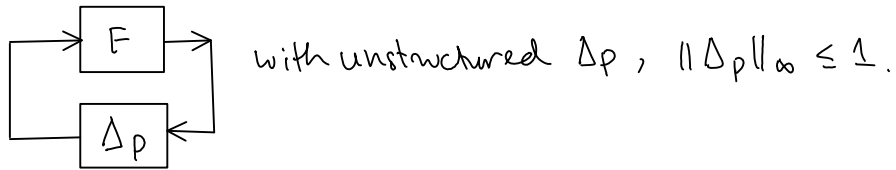


is robustly stable for any allowed  $\Delta$ ,  $\|\Delta\|_{\infty} \leq 1$   
 and any  $\Delta_p$  (unstructured),  $\|\Delta_p\|_{\infty} \leq 1$

$$\hat{\Delta} = \begin{bmatrix} \Delta & 0 \\ 0 & \Delta_p \end{bmatrix}, \|\hat{\Delta}\|_{\infty} \leq 1$$

note that  $\hat{\Delta}$  is structured

Proof: Write RP condition as RS for



Then rewrite  $F$  with the  $N\Delta$  loop, and then regroup the uncertainty.

Observe that, just as in the SISO case, NS and RP implies RS.